## OpenRiskNet

RISK ASSESSMENT E-INFRASTRUCTURE

# Integrating multiple predictions

## ModelRX Case Study

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OpenRiskNet: Open e-Infrastructure to Support Data Sharing, Knowledge Integration and *in silico* Analysis and Modelling in Risk Assessment

Project Number 731075



### Problem

Given multiple evidences make a **consensus** prediction about a blood-brain barrier penetration and estimate its **uncertainty** 

# Approach

**Dempster-Shafer theory** (DST) provides framework for such integrated risk assessment



## The idea behind DST - single evidence

#### Will it rain tomorrow?

- A friend predicts 70 % chance of rain (30 % chance of no-rain)
- We estimate his reliability to 60 %

### What shall we expect?

- Belief of **rain**:  $0.7 \times 0.6 = 0.42$ 

- Belief of **no-rain**:  $0.3 \times 0.6 = 0.18$ 

- **Uncertainty**: 1 - 0.6 = 0.4

## The idea behind DST - single evidence

#### Will it rain tomorrow?

- A friend predicts 70 % chance of rain (30 % chance of no-rain)
- We estimate his <u>reliability to 60 %</u>

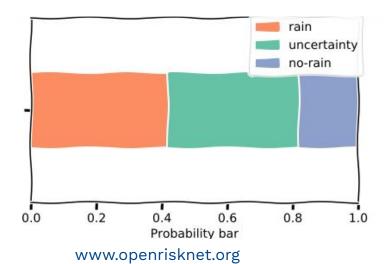
### What shall we expect?

Probability of rain: 0.42 - 0.82

- Probability of **no-rain**: **0.18 - 0.58** 

Uncertainty: 0.4

DST prediction: equivocal





# The idea behind DST - multiple evidences

#### Will it rain tomorrow?

- A friend predicts 70 % chance of rain (60 % reliability)
- Newspaper predicts 60 % chance of rain (80 % reliability)
- Individual predictions are equivocal

### What shall we expect?

- DST provides framework to combine such evidences into a consensus prediction



# The idea behind DST - multiple evidences

#### Will it rain tomorrow?

- A friend predicts 70 % chance of rain (60 % reliability)
- Newspaper predicts 60 % chance of rain (80 % reliability)
- Individual predictions are equivocal

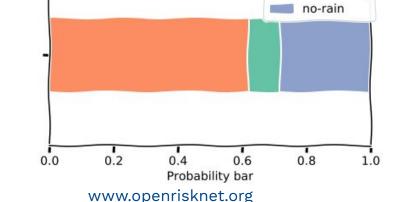
### What shall we expect?

Probability of rain: 0.62 - 0.72

Probability of no-rain: 0.28 - 0.38

- Uncertainty: 0.10

DST prediction: rain



rain

uncertainty



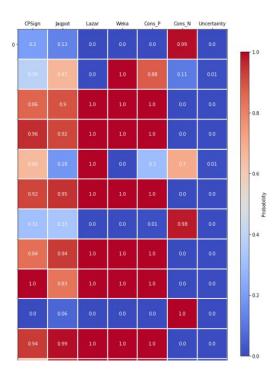
# ModelRX case study

- Predictive models for blood-brain barrier penetration:
   Lazar, Jaqpot, WEKA JGU, Venn-ABERS
- Reliability of each model (cross-validation):
   Positive predictive value (% true positive of all predicted positive)
   Negative predictive value (% true negative of all predicted negative)
- Test set of ~ 400 compounds



# ModelRX case study

- GitHub: <u>OpenRiskNet/workshop/ModelRX/Blood-brain</u>
   <u>barrier Consensus</u>
- Jupyter notebook: consensus-batch-offline.ipynb
- Consensus predictions for 11 compounds of test set





- Consensus prediction using 2 predictive models (Lazar, CPSign)
- GitHub: OpenRiskNet/workshop/ModelRX/Blood-brain barrier Consensus
- Jupyter notebook: consensus-single-web-CL.ipynb

#### Main steps:

- Provide compound structure (SMILES)
- 2. Access web services through REST API (Lazar, CPSign)
- 3. Make a consensus prediction using Python library dst.py



### 1. Provide compound structure (SMILES)

```
smiles = 'COCCC'
```

#### 2. Access the Lazar API

```
r = requests.post(
    url = 'https://lazar.prod.openrisknet.org/model/5ae2dd885f1c2d01323270ee',
    data = {'identifier': smiles},
    headers = {'accept': 'text/csv'}
)

result = json.loads(r.text)

print('Prediction:', result['prediction']['value'])
```

Prediction: penetrating



### 3. Access the CPSign API

```
r = requests.get(
    url = "http://blood-brain-barrier-penetration-cvap-cpsign.prod.openrisknet.org/v1/predict",
    params = {'molecule': smiles},
    headers = {'accept': 'application/json'}
)

result = json.loads(r.text)

result['prediction']

[{'probability': 0.006, 'label': 'non-penetrating'},
    {'probability': 0.994, 'label': 'penetrating'}]
```



#### 4. Make a consensus prediction

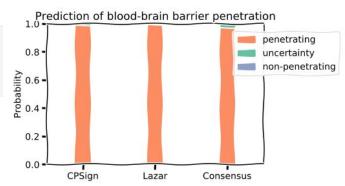
{'CPSign': 0.701, 'Lazar': 0.489}

```
print('Prediction:\t', models_pred)
print('Probability:\t', models_prob)
print('PPV:\t\t', models_ppv)
print('NPV:\t\t', models_npv)

Prediction: {'Lazar': 'P', 'CPSign': 'P'}
Probability: {'Lazar': 1.0, 'CPSign': 0.994}
PPV: {'CPSign': 0.809, 'Lazar': 0.886}
```

```
beliefs, plausibilities, result = dst.predict_Dempster(models_pred, models_prob, models_ppv, models_npv)
print('Beliefs:\t', beliefs)
print('Plausibilities:\t', plausibilities)
print('Result:\t\t', result)

Beliefs: {'P': 0.9775762072502536, 'N': 0.0005557460659842, 'PN': 1.0}
Plausibilities: {'P': 0.9994442539340158, 'N': 0.022423792749746436, 'PN': 1.0}
Result: P
```





NPV:

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